

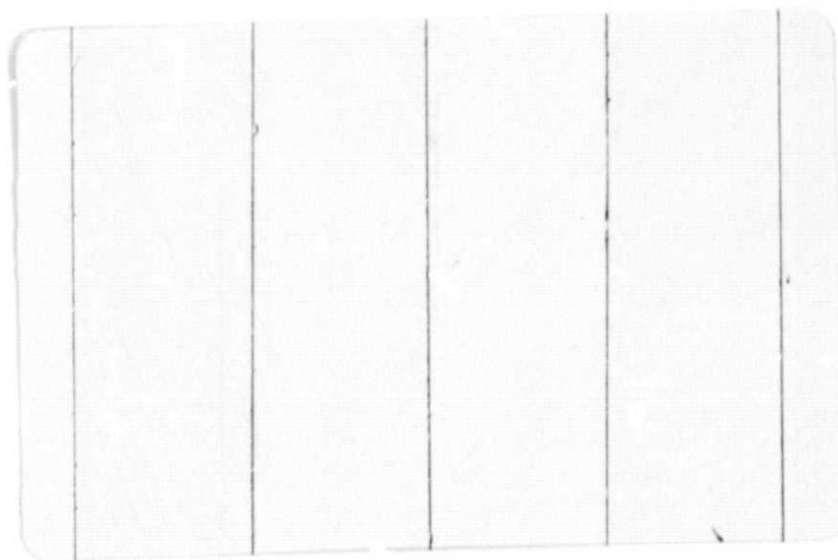
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A Quasi-Newton Approach to
Optimization Problems with
Probability Density Constraints

by

R.A. Tapia and D.L. Van Rooy
Rice University

ABSTRACT:

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Institute for Computer Services and Applications
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I. INTRODUCTION

Consider the optimization problem

$$(1) \quad \begin{aligned} & \min f(x_1, \dots, x_n) ; \text{ subject to } 1 - \sum_{i=1}^n x_i = 0, \\ & x_i \geq 0, \quad i=1, \dots, n \end{aligned}$$

Problems of the form (1) occur frequently in statistical applications. In these applications, the variables x_i usually represent a discrete probability density function and the functional f represents an optimality criterion, e.g., the negative likelihood or negative log likelihood.

Let us make the change of variables

$$(2) \quad x_i = \frac{1}{2} y_i^2, \quad i=1, \dots, n$$

and consider the optimization problem

$$(3) \quad \min \hat{f}(y_1, \dots, y_n) ; \text{ subject to } 1 - \sum_{i=1}^n \frac{1}{2} y_i^2 = 0$$

where

$$(4) \quad \hat{f}(y_1, \dots, y_n) = f\left(\frac{1}{2} y_1^2, \dots, \frac{1}{2} y_n^2\right)$$

Proposition 1.1

If $y = (y_1, \dots, y_n)^T$ solves problem (3), then $x = (\frac{1}{2} y_1^2, \dots, \frac{1}{2} y_n^2)^T$ solves problem (1). Conversely, if $x = (x_1, \dots, x_n)^T$ solves problem (1), then $y = (\sqrt{2x_1}, \dots, \sqrt{2x_n})^T$ solves problem (3).

Proof

The proof is not difficult.

Problem (3), in contrast to problem (1), does not involve inequality constraints. The price one pays for this simplification is that problem (3) will have more critical points than problem (1) and the equality constraint is quadratic instead of linear. We maintain, that according to Proposition 1.1, if one starts sufficiently near the solution, the extraneous critical points will not affect the algorithm. Moreover, when f is nonlinear, the quadratic constraints should not significantly increase the degree of complexity of the problem.

Tapia's approach to quasi-Newton methods for constrained optimization ([1] and [2]) depends heavily on the invertibility of the Hessian of the Lagrangian at the solution. Hence, we are certainly interested in determining the relationship between the invertibility of the Hessian of the Lagrangian for problem (1) and the invertibility of the Hessian of the Lagrangian for problem (3). Toward this end let

$$(5) \quad L(x, \mu, \lambda) = f(x) - \sum_{i=1}^n \mu_i x_i + \lambda(1 - \sum_{i=1}^n x_i)$$

and

$$(6) \quad \hat{L}(y, \lambda) = \hat{f}(y) + \lambda(1 - \sum_{i=1}^n \frac{1}{2} y_i^2)$$

we use the notation

$$\Lambda = (\lambda, \dots, \lambda)^T,$$

$$Y = \text{diag}(y_1, \dots, y_n)$$

and

$$X = \text{diag}(x_1, \dots, x_n)$$

Straightforward calculations show that

$$(7) \quad \nabla_y \hat{L}(y, \lambda) = Y \left(\nabla f(x) - \lambda \right) ,$$

$$(8) \quad \nabla_y^2 L(y, \lambda) = Y \nabla^2 f(x) Y + \text{diag} \left(\nabla f(x) - \lambda \right) ,$$

$$(9) \quad \nabla_x L(x, \mu, \lambda) = \nabla f(x) - \lambda - \mu ,$$

and

$$(10) \quad \nabla_x^2 L(x, \mu, \lambda) = \nabla^2 f(x) .$$

The first order necessary conditions for problem (1)

are

$$(11) \quad \begin{aligned} \nabla f(x) - \lambda - \mu &= 0 \\ 1 - \sum_{i=1}^n x_i &= 0 \\ \mu_i &\geq 0 \\ x_i &\geq 0 \\ \mu_i x_i &= 0 , \quad i=1, \dots, n ; \end{aligned}$$

while the first order necessary conditions for problem (3) are

$$(12) \quad \begin{aligned} Y \left(\nabla f(x) - \lambda \right) &= 0 \\ 1 - \sum_{i=1}^n \frac{1}{2} y_i^2 &= 0 \end{aligned}$$

Recall that strict complementarity means that in (11) we do not have $\mu_i = x_i = 0$ for any i .

Proposition 1.2

Let (x, u, λ) be a solution of problem (1) and (y, λ) the corresponding solution of problem (3). Then

(i) we have strict complementarity and $\nabla^2 f(x)$ is positive definite $\iff \nabla_y^2 \hat{L}(y, \lambda)$ is positive definite

(ii) we have strict complementarity and $\nabla^2 f(x)$ is invertible $\iff \nabla_y^2 L(y, \lambda)$ is invertible.

Proof

Again the proof is not difficult. One merely works with (8), (10), (11) and (12).

Proposition 1.2 is very satisfying and warns us that in the algorithm we must guard against the loss of strict complementarity. This will be accomplished by working with the variable

$$(13) \quad x_i = \frac{1}{E_i} y_i^{E_i}$$

where $E_i = 2$ if $|x_i| + |\mu_i| \neq 0$ and $E_i = 1$ if $|x_i| + |\mu_i| = 0$. Observe that, when $E_i = 1$ in (13), we have effectively ignored the nonnegativity constraint on x_i .

II. THE QUASI-NEWTON ALGORITHM

The algorithm we are about to present is a special case of the approach given by Tapia in [2] applied to problem (3). The reader is referred to that paper for a better understanding of the theory and algorithm. In describing an iterative method, it is convenient to denote the present iterate by x and the subsequent iterate by \bar{x} and similarly for other variables. Also, we let I

denote the identity matrix, $\langle \cdot, \cdot \rangle$ denotes the Euclidean inner product, $e_1 = (1, 0, \dots, 0)^T, \dots, e_n = (0, \dots, 1)^T$, (a_{ij}) denotes the matrix whose (i, j) -th element is a_{ij} and b_i or $(b)_i$ denotes the i -th element of the vector b .

Consider the constrained optimization problem

$$(14) \quad \min f(x), \quad \text{subject to } g(x) = 0;$$

where $f, g: \mathbb{R}^n \rightarrow \mathbb{R}^1$. We first describe the quasi-Newton method BFGS given in [2] for problem (14). Let

$$(15) \quad L(x, \lambda) = f(x) + \lambda g(x)$$

Quasi-Newton BFGS

Step 1: (Initialization) Determine x and B .

Step 2: (Update x and λ) Let

$$\bar{\lambda} = \frac{g(x) - \langle \nabla g(x), B^{-1} \nabla f(x) \rangle}{\langle \nabla g(x), B^{-1} \nabla g(x) \rangle}$$

$$\bar{x} = x - B^{-1} \nabla_x L(x, \bar{\lambda}).$$

Step 3: (Update approximate Hessian)

$$\bar{B} = \left(B + \frac{y y^T}{\langle y, s \rangle} - \frac{B s s^T B}{\langle s, B s \rangle} \right)$$

where

$$s = \bar{x} - x$$

and

$$y = \nabla_x L(\bar{x}, \bar{\lambda}) - \nabla_x L(x, \bar{\lambda}).$$

Step 4: GO TO STEP 2.

In implementing this algorithm, one would have to make provisions for output and a stopping criterion. In [2], Tapia has demonstrated that the above algorithm is locally superlinearly convergent.

Remark 1:

If the initial B depended on λ , we would make an initial approximation of λ according to the projection formula (see [2])

$$(16) \quad \lambda = - \frac{\langle \nabla g(x), \nabla f(x) \rangle}{\langle \nabla g(x), \nabla g(x) \rangle}$$

Remark 2:

In updating B in Step 3, it is advantageous to store B factored according to a Cholesky decomposition and then merely update these factors. Moreover, B can be slightly modified so that the resultant matrix is always positive definite. For a detailed description of these modifications, the reader is referred to Gill et al [3], Van Rooy et al [4], Fletcher and Powell [5] and Van Rooy [6]. The quantities $B^{-1} \nabla f(x)$ and $B^{-1} \nabla g(x)$ in Step 2 are then obtained in the standard back-substitution manner.

We now apply the above algorithm for problem (3) with the additional feature of obtaining a discrete Newton approximation to the Hessian every M iterations (where M is an input parameter).

Define the following functions

$$(17) \quad x_i = \frac{y_i E_i}{E_i}$$

$$(18) \quad \nabla g_i = - \left[2 - E_i + (E_i - 1) y_i \right], \\ i=1, \dots, n$$

and

$$(19) \quad g = 1 - \sum_{i=1}^n x_i$$

The quantity ∇g is merely the gradient of g with respect to y and (13) and (17) are the same.

Quasi-Newton Algorithm for Problem (3)

Step 1: (Initialization)

Read	CRC	(constraint rejection criterion)
	h	(discretization step)
	ϵ	(stopping criterion)
	DI	(descent indicator)
	IOUTPUT	(output indicator)
	M	(discrete Hessian indicator)
	MAX	(maximum number of iterations)

Let

$$E_i = 2$$

$$y_i = \sqrt{\frac{2}{n}}$$

$$\lambda = \langle \nabla f(x), x \rangle$$

$$B = I$$

Step 2: Output according to IOUTPUT,
check stopping criterion

IF $(\|\nabla_y L(y, \lambda)\| + g^2 < \epsilon$ or
 ITERATION > MAX) stop

Step 3: Calculate discrete Hessian according to M

$$B = \left(\frac{\nabla_y \hat{L}(y + h e_j, \lambda)_i - \nabla_y \hat{L}(y, \lambda)_i}{h} \right)$$

Calculate modified Cholesky decomposition of B .

Step 4: Calculate multiplier correction

$$\Delta \lambda = \frac{g - \langle \nabla g, B^{-1} \nabla_y \hat{L}(y, \lambda) \rangle}{\langle \nabla g, B^{-1} \nabla g \rangle}$$

Step 5: Calculate y-variable correction

$$\Delta y = - B^{-1} \nabla_y \hat{L}(y, \lambda) - \Delta \lambda B^{-1} \nabla g$$

Step 6: Update y according to descent indicator

$$\bar{\lambda} = \lambda + \alpha \Delta \lambda$$

$$\bar{y} = y + \alpha \Delta y$$

where $\alpha = 1$ if descent is not desired, other-
 wise α is chosen according to the descent
 principle on $\|\nabla_y \hat{L}(y, -\lambda)\|^2 + g^2$
 (see [2]).

Step 7: Update approximate Hessian according to BFGS
 and modified Cholesky decomposition

$$\bar{B} = B + \frac{Dy Dy^T}{\langle Dy, Ds \rangle} - \frac{B Ds Ds^T B}{\langle Ds, B Ds \rangle}$$

where $Ds = \bar{x} - x$ and

$$Dy = \nabla_y \hat{L}(\bar{y}, \bar{\lambda}) - \nabla_y \hat{L}(y, \bar{\lambda})$$

Step 8: Update constraint rejection indicators

$$T_0 = a_1 \left[\|\nabla_y \hat{L}(\bar{y}, \bar{\lambda})\|^2 + \left(1 - \sum_{i=1}^n x_i\right)^2 \right]^{a_2},$$

$$T_1 = \min \left(\frac{1}{4n}, T_0 \right),$$

$$T_2 = \min (CRC, T_0)$$

$$\text{If } (|\bar{x}_i| < T_1 \text{ and } \left| \frac{\partial f}{\partial x_i} - \lambda \right| < T_2),$$

let $E_i = 1$

If E_i has changed, reset $\bar{y}_i \quad i=1, \dots, n$

Step 9: GO TO STEP 2 .

III. USER'S GUIDE

The above algorithm has been implemented in a set of FORTRAN subroutines whose entry point is the subroutine CONOPT. In this section, we shall describe how to use this program, provide a description of the workings of the program, and illustrate its use with examples. A listing of the program is given in the appendix. This program runs on an IBM 370/155 using the FORTRAN G, G1, or H compilers. All computations are done in double precision. Unit 6 is used for output, if requested.

To utilize this program, the user supplies several parameters and two subprograms which calculate the function value and its gradient at a specified point. The program will then attempt to minimize the function subject to the constraints and return the argument at the minimum.

The calling sequence of the program is `CALL CONOPT (N, DELF, FUNCT, X, WKA, MITER, STC, CRC, IDN, DH, IOUT, DSNT, A, RSTRT, USELAM, XLAM)`.

These parameters are described below:

- N - integer variable indicating the dimension of the problem.
- DELF - a user supplied subroutine which calculates the gradient of the function. The calling sequence is `CALL DELF (DF, X, N)`, where X is the N dimensional argument and DELF returns the gradient in the vector DF. Note that DF and X are both double precision variables.

FUNCT - a user supplied double precision function subprogram which calculates the function value at a specified point. The calling sequence is $F = \text{FUNCT}(X, N)$ where X is the (double precision) argument of length N . Though the function value is not required by the algorithm, **CONOPT** will output it in the convergence test step if the user requests output. (It is called only when output is requested.) If, for any reason, the user is not interested in this value, he may just supply a function subprogram that sets the function value to zero and returns.

[N.B. The two programs **DELF** and **FUNCT** must be declared external in the calling program. The two programs may have other entry points, with different arguments, referred to by the user's calling program prior to his calling **CONOPT** in order to make other variables or their addresses available to either or these subprograms.]

X - on output, this will contain the N -dimensional double precision argument last calculated by the program (at the minimum if the program converged). On input, the program will set each element of X to the value $1/N$, unless the user has specified restart (see description of the variable **RSTRT** below), in which case

the supplied values of X will be used as the starting point.

- WKA - a double precision working storage array of length $\geq \frac{1}{2} N^2 + (23N + 7) / 2$. If a restart has been specified, the first N (2-byte integer) locations should contain the desired constraint rejection indicators. However, if these variables do not contain either 1's or 2's, the program will set them to 2. The final constraint rejection indicators will be in these locations.
- MITER - an integer variable indicating the maximum number of iterations to perform. If the user sets this to 0, the program will reset it to $\max(30, 2 * N)$.
- STC - a real variable specifying the stopping criterion on the norm squared of the gradient of the Lagrangian plus the constraint, i.e. $\| \nabla L \|^2 + \left(1 - \sum_{i=1}^n x_i \right)^2$. If 0 is specified, the program will set $STC = 10^{-8}$.
- CRC - a real variable specifying the constraint rejection criterion. See the algorithm description section for a precise definition of its use. If 0 is specified, the program will set $CRC = 10^{-2}$.
- IDN - an integer variable specifying the frequency of calculation of the discrete approximation to the Hessian, i.e., this finite difference Hessian is calculated every IDN iterations. If 0 is specified, this approximation is never used.

If a negative value is specified, the program sets $IDN = \min(10, N)$.

- DH - a real variable specifying the step size to be used in calculating the finite difference Hessian (see the algorithm description for a detailed definition). If 0 is specified, the program will set $DH = 10^{-2}$.
- IOUT - an integer variable specifying the frequency of output; i.e., the program will output various computed quantities every IOUT iterations. If 0 is specified, no output will be produced. If a negative value is specified, the program sets $IOUT = 1$.
- DSNT - a one-byte logical variable indicating whether the program should implement descent in its search at each iteration.
- A - a two-word real vector containing respectively the multiplier and exponent for DNORM

$$\left(= \|\nabla L\|^2 + \left(1 - \sum_{i=1}^n x_i\right)^2 \right)$$
for use in the constraint rejection test; i.e., the appropriate quantity is $A(1) * DNORM * A(2)$. The defaults for $A(1)$ and $A(2)$ are 1. and .2 respectively. If the user wishes to set $A(2) = 0.$, he should set $A(1)$ to the negative of what he wants for $A(1)$ and then a value of zero will be used for $A(2)$.

RSTRT - a one-byte logical variable indicating whether the user is supplying an initial value for X (as in a restart case). If RSTRT = .TRUE., the first $2N$ bytes of WKA should contain the vector ICR (see description of WKS and the algorithm description—the E_i 's).

USELAM - a one-byte logical variable used if RESTR = .TRUE. to indicate if the user is supplying a value for XLAM. If RESTR = .TRUE. and USELAM = .FALSE., the value $\langle \nabla f, x \rangle$ is used for XLAM.

XLAM - the Langrange multiplier for the equality constraint. Used on input only if USELAM = RSTRT = .TRUE.

The default values for the above parameters have been chosen so that many problems will converge; thus, the only input parameters with which the user need concern himself in many cases is N , IOUT, DSNT and RSTRT (usually = .FALSE.).

The subprograms used and a brief description of their function is given below:

CONOPT - main driver subroutine

DIMS - computes base addresses for various working arrays used throughout the calculations.

INPT - sets default values for parameters and outputs the parameters used.

INIT initializes y , x , ICR, H and λ . Also fetches values of ∇f and ∇L .

- CHKSTP - checks the stopping criterion and outputs quantities if requested.
- DHESS - calculates a finite difference approximation to the Hessian and its modified Cholesky decomposition, if requested. The Hessian is forced to be positive definite.
- DELAMY - computes the update terms $\nabla \lambda$ and ∇y .
- NEWY - updates y and λ based optionally on descent of $\|\nabla L\|^2 + \left(1 - \sum_{i=1}^n x_i\right)^2$.
- UPDTH - updates the approximate Hessian using the Broyden-Fletcher-Goldfarb-Shannon (BFGS) update technique. (Really the MCD of it is updated.) Here again the Hessian is forced to be positive definite.
- CONSTR - updates the constraint rejection indicators and, if necessary, y and ∇L .
- GRADL - computes ∇L given ∇f , λ , y , and ICR.
- SUM - a function subprogram that performs various utility functions $\left(\sum_{i=1}^n x_i, \sum_{i=1}^n x_i^2, \sum_{i=1}^n x_i y_i\right)$.
- MCHLSK - computes the modified Cholesky decomposition (MCD) of a matrix. If the matrix is not positive definite, the MCD is modified to represent a matrix that is (see e.g. ref 4).
- PDSOLV - solves $Ax = y$ where A contains the MCD of a matrix.

COMPT - computes a rank one update of an MCD using the composite-t method (see refs. 5 and 6).

LDMUL - multiply a vector by a matrix whose MCD is supplied.

Examples:

The following quadratic programming (i.e., $\min \|Ax - b\|^2$ s. t. $x_i \geq 0$ and $\sum x_i = 1$) examples were run using all the default options:

$$1. \quad A = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \quad b = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$x = (.11112108, .44443959, .44443959)^T$$

converged in 5 iterations. Exact answer = $(1/9, 4/9, 4/9)^T$

The robustness of the algorithm is exemplified by the following two examples for which the algorithm, theoretically, should not converge since the Hessian is singular at the solution.

$$2. \quad A = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 2 & 1 & 2 \end{pmatrix} \quad b = \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix}$$

$$x = (.50000019, 0., .50000004)^T$$

converged in 14 iterations. Exact answer = $(\frac{1}{2}, 0, \frac{1}{2})^T$.

The constraint for x_2 is not active in this case.

$$3. \quad A = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} \quad b = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

$$x = (0, 0, 1)^T$$

congerged in 11 iterations. The constraints for x_1 and x_2 are not active in this case.

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APPENDIX

Program Listings of CONOPT

LEVEL 21.8 (JUN 74)

DS/360 FORTRAN H

COMPILER OPTIONS -

NAME= MAIN,OPT=00,LINECNT=60,SIZE=0000K,
 SOURCE,EBCDIC,NOLIST,NODECK,LOAD,MAP,NOEDIT,ID,XREF
 SUBROUTINE CONOPT(N,DELF,FUNCT,X,WKA,TMITER,TSTC,TCRC,TIDN,TDH,
 1 TIOUT,TDSNT,A,RSTRT,USELAM,XLAM)
 IMPLICIT REAL*8 (A-H,O-Z)

ISN 0002

ISN 0003

THIS ROUTINE MINIMIZES F(X) S.T. $X(1) \cdot GE .0$ & $SUM(X(1)) = 1$.
 FOR A DETAILED DESCRIPTION OF THE PROGRAM SEE ICOSA TECHNICAL
 REPORT NO. 275-025-029, RICE UNIV., HOUSTON, TEX, JUNE, 1976.

FUNCT IS A FUNCTION WHICH RETURNS F(X), ITS ARGUMENTS ARE X,N
 DELF IS A SUBROUTINE WHICH COMPUTES THE GRADIENT OF F(X). ITS
 ARGUMENTS ARE DF (THE GRAD), X, & N

DELF & FUNCT MUST BE DECLARED EXTERNAL IN THE MAIN PROGRAM.

WKA IS A WORK AREA OF LENGTH .GE. $.5 * N * (2 + (23 * N + 7) / 2)$

X ON RETURN IS THE ARGUMENT AT THE MINIMUM

N IS THE DIMENSION OF X

A(1) & A(2) ARE USED IN THE CONSTRAINT REJECTION TEST. AS
 A(1)*DNORM**A(2) DEFAULTS ARE A(1)=1. & A(2)=.2. IF THE USER
 WISHES A(2) TO =0, HE SHOULD SET A(1) TO THE NEGATIVE OF WHAT HE
 WANTS FOR A(1).

RSTRT - A LOGICAL VARIABLE INDICATING WHETHER THE USER IS
 RESTARTING. IF SO, THE X'S MUST BE SUPPLIED & THE CONSTRAINT
 REJECTION INDICATORS AS INTEGER*2 VARIABLES SHOULD BE IN THE FIRST
 N LOCATIONS OF WKA

USELAM - A LOGICAL VARIABLE INDICATING WHETHER OR NOT THE USER
 IS SUPPLYING A VALUE FOR XLAM IN THE CASE OF A RESTART.

XLAM - THE LAGRANGE MULTIPLIER FOR THE EQUALITY CONSTRAINT.

REAL*8 X(N),WKA(1)

REAL*4 TSTC,TCRC,TDH

REAL*4 STC,CRC,DH

REAL*4 A(2)

INTEGER*4 TMITER,TIDN,TIOUT

LOGICAL*1 DSNT,

COMMON /INPUT/ MITER,STC,CRC,IDN,DH,IOUT,DSNT

EXTERNAL DELF,FUNCT

COMMON /ARANGE/ IICR,IY,IH,IDF,IGL, IGL2,IDY,IWK,IGY,IYSAV,

1 IHGL,IHGY,IDLY,IDS

COMPUTE THE STARTING ADDRESSES OF VARIOUS WORKING ARRAYS.

CALL DIMS(N)

SET DEFAULT VALUES & OUTPUT (STEP 1)

CALL INPT(TMITER,TSTC,TCRC,TIDN, TDH,TIOUT,TDSNT,N,A,RSTRT)

INITIALIZE VARIABLES

CALL INIT(X,WKA(IDF),WKA(IY),WKA(IH),WKA(IGL),WKA(IICR),XLAM,

1 NITER,N,DELF,RSTRT,USELAM)

STEP 2 - CHECK STOPPING CRITERION & OUTPUT IF REQUESTED

100 CALL CHKSTP (X,N,G,DNORM,NITER,XLAM,WKA(IGL),

FUNCT, 2200)

(1)

ISN 0004

ISN 0005

ISN 0006

ISN 0007

ISN 0008

ISN 0009

ISN 0010

ISN 0011

ISN 0012

ISN 0013

ISN 0014

ISN 0015

ISN 0016


```

C C
ISN 0017 STEP 3 ~ CALCULATE DISCRETE HESSIAN (OPTIONAL)
C C
C C CALL DHESS ( NITER,WKA(IH), WKA(IWK),WKA(IGL),
C C 1 WKA(IGL2), DELF, XLAM,WKA(IICR),WKA(IY),N,X)
C C
C C STEPS 4 & 5 ~ CALCULATE MULTIPLIER & Y-VARIABLE CORRECTIONS
C C
C C CALL DELAMY (WKA(IGY),WKA(IH),WKA(IGL),N,WKA(IHGL),WKA(IHGY),
C C 1 DLAM,WKA(IDLY),G,WKA(IY),WKA(IICR))
C C
C C STEP 6 ~ CALCULATE NEW Y BASED ON (OPTIONAL) DXSCENT
C C
C C CALL NEWY (DNORM,XLAM,WKA(IY),N,DLAM,WKA(IDLY),G,NITER,WKA(IICR),
C C 1 WKA(IYSAV),X,WKA(IDF),DELF,WKA(IDY),WKA(IGL))
C C
C C STEP 7 ~ UPDATE APPROXIMATE INVERSE HESSIAN
C C
C C CALL UPDTH(NITER,WKA(IYSAV),N, WKA(IGL),XLAM,WKA(IICR),
C C 1 WKA(IY),WKA(IH), WKA(IGL2),WKA(IDY),WKA(IDS),WKA(IWK))
C C
C C STEP 8 ~ UPDATE CONSTRAINT REJECTION INDICATORS
C C
C C CALL CONSTR(WKA(IDF),X,N,WKA(IICR),XLAM,DELF,WKA(IGL),WKA(IY),
C C 1 WKA(IYSAV),DNORM,A)
C C
C C GO TO 100
ISN 0021 200 RETURN
ISN 0022 END
ISN 0023
ISN 0024

```

```

      COMPILER OPTIONS = NAME= MAIN,OPT=00,LINECNT=60,SIZE=0000K,
      SOURCE,EBCDIC,NOLIST,NODECK,LOAD,MAP,NOEDIT,ID,XREF
      SUBROUTINE DIMS(N)
      C
      C COMPUTE STARTING POINTS OF WORKING ARRAYS USED.  THE ORDER IS ICR,
      C Y,H, DF,GL,GL2,DY, & WK
      C
      INTEGER*4 IDM(8)
      COMMON /ARANGE/ IDM,IGY,IYSAV,IHGL,IHGY,IDLY,IDS
      IDM(1)=1
      IDM(2)=(N+1)/2
      IDM(3)=N
      IDM(4)=N*(N+1)/2
      IDM(5)=N
      IDM(6)=N
      IDM(7)=N
      IDM(8)=N
      DO 10 I=2,8
      10 IDM(I)=IDM(1)+IDM(I-1)
      IGY=IDM(8)
      IYSAV=IGY+N
      IHGL=IYSAV+N
      IHGY=IHGL+N
      IDLY=IHGY+N
      IDS=IDLY+N
      RETURN
      END

```

COMPILER OPTIONS - NAME= MAIN,OPT=00,LINECNT=60,SIZE=0000K,
SOURCE,EBCDIC,NOLIST,NODECK,LOAD,MAP,NODEDIT,ID,XREF
SUBROUTINE INPT (TMITER,TSTC,TCRC,TIDN,TDH,TIOUT,TDSENT,N,A,RSTRT)

ISN 0002

C
C
C

SET DEFAULT VALUES FOR PARAMETERS USED.

```

LOGICAL*1 DSNT,TDSENT,RSTRT
INTEGER*4 TMITER,TIDN,TIOUT
REAL*4 A(2)
REAL*4 TSTC,TCRC,TDH
REAL*4 STC,CRC,DH
COMMON /INPUT/ MITER,STC,CRC,IDN,DH,IOUT,DSNT
NAMELIST /IN/ MITER,STC,CRC,IDN,DH,IOUT,DSNT
EPS=1.E-20
MITER=MAX0(30,2*N)
IF (TMITER.GT.0) MITER=TMITER
STC=1.E-8
IF (TSTC.GT.EPS) STC=TSTC
CRC=.01
IF (TCRC.GT.EPS) CRC=TCRC
IDN=MIN0(10,N)
IF (TIDN.GE.0) IDN=TIDN
DH=.01
IF (TDH.GT.EPS) DH=TDH
IOUT=1
IF (TIOUT.GE.0) IOUT=TIOUT
DSNT=TDSENT
IF (A(2).LE.EPS) A(2)=.2
IF (A(1).LT.0.) A(2)=0.
A(1)=ABS(A(1))
IF (A(1).LE.EPS) A(1)=1.
WRITE (6,11)
11 FORMAT (',1 PARAMETERS USED')
WRITE (6,12) A
12 FORMAT (', A(1)=',G16.8,' A(2)=',G16.8)
13 WRITE (6,13) RSTRT
13 FORMAT (', RSTRT=',L8)
RETURN
END

```

COMPILER OPTIONS -

NAME= MAIN,OPT=00,LINECNT=60,SIZE=0000K,
 SOURCE,EBCDIC,NOLIST,NODECK,LOAD,MAP,NOEDIT,ID,XREF
 SUBROUTINE INIT(X,DF,Y,H,GL,ICR,XLAM,NITER,N,D,DEL,RSRT,USELAM)
 IMPLICIT REAL*8 (A-H,O-Z)

THIS ROUTINE INITIALIZES SEVERAL VARIABLES USED IN THE OPTIMIZATION

ISN 0002
 ISN 0003

C
 C

ISN 0004
 ISN 0005
 ISN 0006
 ISN 0007
 ISN 0008
 ISN 0010
 ISN 0011
 ISN 0012
 ISN 0013
 ISN 0014
 ISN 0015
 ISN 0016
 ISN 0017
 ISN 0018
 ISN 0020
 ISN 0021
 ISN 0022
 ISN 0023

LOGICAL*1 USELAM
 LOGICAL*1 RSRT
 INTEGER*2 ICR(N)
 REAL*8 X(N),DF(N),Y(N),GL(N),H(1)
 IF (RSRT) GO TO 5
 SN=SQRT(2./N)
 DO 10 I=1,N
 ICR(I)=2
 Y(I)=SN
 10 X(I)=1./N
 GO TO 15
 5 DO 40 I=1,N
 Y(I)=X(I)
 IF (ICR(I).EQ.1) GO TO 40
 ICR(I)=2
 Y(I)=DSQRT(2.*X(I))
 40 CONTINUE
 15 CONTINUE

C
 C
 C

DF IS THE GRADIENT OF F

ISN 0024

CALL DELF(DF,X,N)

C
 C
 C

SET H TO THE IDENTITY MATRIX

ISN 0025
 ISN 0026
 ISN 0027
 ISN 0028
 ISN 0029
 ISN 0030
 ISN 0031
 ISN 0032
 ISN 0034

N2=N*(N+1)/2
 DO 20 K=1,N2
 20 H(K)=0.
 I=0
 DO 30 K=1,N
 I=I+K
 30 H(I)=1.
 IF (.NOT.USELAM) XLAM=DOT(DF,X,N)
 NITER=0

C
 C
 C

GL IS THE GRADIENT OF THE LAGRANGIAN

ISN 0035
 ISN 0036
 ISN 0037

CALL GRADL(GL,DF,XLAM,ICR,Y,N)
 RETURN
 END

```

      COMPILER OPTIONS = NAME= MAIN,OPT=00,LINECNT=60,SIZE=0000K,
      SOURCE,EBCDIC,NOLIST,NODECK,LOAD,MAP,NOEDIT,;D,XREF
      SUBROUTINE CHKSTP(X,N,G,DNORM,NITER,XLAM,GL,
1*)
      IMPLICIT REAL*8 (A-H,O-Z)
      REAL*8 SUM,SUM2
      REAL*8 X(N),GL(N)
      REAL*4 STC,CRC,DH
      COMMON /INPUT/ MITER,STC,CRC,IDN,DH,IOUT
      EVALUATE THE NORM OF GRAD(L) + THE NORM OF THE EQUALITY CONSTRAINT, G.
C
C
      G=1.D0-SUM(X,N)
      DNORM=G*G+SUM2(GL,N)
      IF (DNORM-GE.STC-AND.NITER.LT.MITER) GO TO 15
      IF (IOUT.EQ.0) RETURN 1
      F=FUNCT(X,N)
      WRITE (6,11) NITER,DNORM,G,XLAM,F,(X(I),I=1,N)
11 FORMAT (//,1 NITER,DNORM,G,XLAM,F,8 X,110,4G16.8,(/8G16.8))
      RETURN 1
C
      INTERMEDIATE OUTPUT IF DESIRED
C
C
15 IF (IOUT.EQ.0) GO TO 25
   IF (NITER/IOUT*IDN-NE.NITER) GO TO 25
   F=FUNCT(X,N)
   WRITE (6,11) NITER,DNORM,G,XLAM,F,(X(I),I=1,N)
25 NITER=NITER+1
   RETURN
   END

```



```

      COMPILER OPTIONS = NAME= MAIN,OPT=00,LINECNT=60,SIZE=0000K,
      SOURCE,EBCDIC,NOLIST,NODECK,LOAD,MAP,NODEIT,ID,XREF
      SUBROUTINE DELAMY(GY,H,GL,N,HGL,HGY,DLAM,DLY,G,Y,ICR)
      IMPLICIT REAL*8 (A-H,O-Z)
      C
      C      COMPUTE DELTA=LAMBDA & DELTA=Y
      C
      REAL*8 GY(N),H(1), GL(N),HGL(N),HGY(N),DLY(N),Y(N)
      INTEGER*2 ICR(N)
      DO 10 I=1,N
      GY(I)=Y(I)
      10 IF (ICR(I).NE.2) GY(I)=1.
      CALL PDSOLV(H,HGL,GL,N)
      CALL PDSOLV(H,HGY,GY,N)
      DLAM=(G-DDOT(GY,HGL,N))/DOT(GY,HGY,N)
      DO 20 I=1,N
      20 DLY(I)=HGL(I)-DLAM*HGY(I)
      RETURN
      END

```

ISN 0002
ISN 0003

C
C
C

ISN 0004
ISN 0005
ISN 0006
ISN 0007
ISN 0008
ISN 0010
ISN 0011
ISN 0012
ISN 0013
ISN 0014
ISN 0015
ISN 0016

COMPILER OPTIONS -

NAME= MAIN,OPT=00,LINECNT=50,SIZE=0000X,
SOURCE,EBCDIC,NOLIST,NODECK,LOAD,MAP,NOEDIT,ID,XREF
SUBROUTINE NEWY (DNORM,XLAM,Y,N,DLAM,DELY,G,NITER,ICR,YSAY,X,
1 DF,DELF,DFSAY,GL)
IMPLICIT REAL*8 (A-H,O-Z)

ISN 0002

ISN 0003

C
C
C

COMPUTE A NEW VALUE OF Y BASED (OPTIONALLY) ON THE DESCENT OF THE
NORM OF GRAD(L) + NORM OF G.

ISN 0004

ISN 0005

ISN 0006

ISN 0007

ISN 0008

ISN 0009

ISN 0010

ISN 0011

ISN 0012

ISN 0013

ISN 0014

ISN 0015

ISN 0016

ISN 0017

ISN 0018

ISN 0019

ISN 0020

ISN 0021

ISN 0022

ISN 0023

ISN 0025

ISN 0026

ISN 0027

REAL*8 SUM2

REAL*8 SUM

LOGICAL*1 DSNT,DSNT2

REAL*8 GL(N),DFSAY(N)

REAL*8 Y(N),DELY(N),YSAY(N),X(N),DF(N)

REAL*4 STC,CRC,DH

INTEGER*2 ICR(N)

COMMON /INPUT/ NITER,STC,CRC,IDN,DH,IOUT,DSNT

USNT2=DSNT

ALP=1.

DNSAVE=DNORM

XLSAV=XLAM

DO 10 I=1,N

DFSAY(I)=DF(I)

YSAY(I)=Y(I)

DO 20 I=1,10

XLAM=XLSAV+ALP*DLAM

DO 30 J=1,N

Y(J)=YSAY(J)+ALP*DELY(J)

IF (ICR(J).EQ.1.AND.Y(J).LT.0.) Y(J)=0.

X(J)=Y(J)*ICR(J)/ICR(J)

30 CONTINUE

CALL DELF(DF,X,N)

C
C

DESCEND IF DESIRED

G=1.D0-SUM(X,N)

CALL GRADL(GL,DF,XLAM,ICR,Y,N)

IF (.NOT.DSNT2) RETURN

DNORM=G*G+SUM2(GL,N)

IF (DNORM.LT.DNSAVE) RETURN

20 ALP=ALP/2.

IF (ALP.LT.0.) GO TO 25

TRY OTHER DIRECTION

C
C

ALP=-1.

GO TO 15

25 WRITE (6,11) NITER

11 FORMAT (' *ON ITER',IS,' DESCENT COULD NOT BE IMPLEMENTED. STEP LE

NGTH SET TO 1.')

ALP=1.

DSNT2=.FALSE.

GO TO 15

RETURN

END

ISN 0038

ISN 0039

ISN 0040

ISN 0041

ISN 0042

ISN 0043

ISN 0044

ISN 0045

ISN 0046


```

COMPILER OPTIONS = NAME= MAIN,OPT=00,LINECNT=60,SIZE=0000K,
SOURCE,EBCDIC,NOLIST,NODECK,LOAD,MAP,NOEDIT,ID,XREF
SUBROUTINE CONSTR(DF,X,N,ICR,XLAM,DELF,GL,Y, YSAV,DNORM,A)
IMPLICIT REAL*8 (A-H,O-Z)

```

CHECK FOR REDUNDANT CONSTRAINTS & RESET CONSTRAINT INDICATORS

```

ISN 0004 INTEGER*2 ICR(N)
ISN 0005 REAL*8 YSAV(N)
ISN 0006 REAL*8 DF(N),X(N)
ISN 0007 REAL*8 GL(N),Y(N)
ISN 0008 REAL*4 STC,CRC,DH
ISN 0009 REAL*4 A(2)
ISN 0010 LOGICAL*1 IND
ISN 0011 COMMON /INPUT/ MITER,STC,CRC,INDN,DH,IOUT
ISN 0012 XML=1.
ISN 0013 SN=SQRT(FLOAT(N))
ISN 0014 EPS=.25
ISN 0015 IND=.FALSE.

```

TESTING & RESETTNG LOOP

```

0016 ISN 0016 DO 10 I=1,N
0017 ISN 0017 T=DF(I)=XLAM
0018 ISN 0018 TS=A(I)*DNORM**A(2)
0019 ISN 0019 CR=EPS/N
0020 ISN 0020 CR=DMINI(CR,TS)
0021 ISN 0021 IF (DABS(X(I)).GE.CR) GO TO 12
0022 ISN 0022 CR=DMINI(CR,TS)
0023 ISN 0023 IF (DABS(T).GE.CR) GO TO 12
0024 ISN 0024 IF (ICR(I).NE.1) IND=.TRUE.
0025 ISN 0025 ICR(I)=1
0026 ISN 0026 GO TO 10
0027 ISN 0027 12 CONTINUE
0028 ISN 0028 IF (ICR(I).NE.2) IND=.TRUE.
0029 ISN 0029 ICR(I)=2
0030 ISN 0030 10 CONTINUE
0031 ISN 0031 15 CONTINUE
0032 ISN 0032 IF (IOUT.EQ.0) GO TO 25
0033 ISN 0033 IF (IOUT*(NITER/IOUT).EQ.NITER) WRITE (6,21) (ICR(I),I=1,N)
0034 ISN 0034 21 FORMAT (' ICR',20I5)
0035 ISN 0035 25 CONTINUE
0036 ISN 0036 IF (.NOT.IND) RETURN
0037 ISN 0037 DO 30 I=1,N
0038 ISN 0038 Y(I)=X(I)
0039 ISN 0039 IF (ICR(I).NE.2) GO TO 30
0040 ISN 0040 Y(I)=0.
0041 ISN 0041 IF (X(I).GT.0.) Y(I)=DSQRT(2.*X(I))
0042 ISN 0042 30 CONTINUE
0043 ISN 0043 CALL GRADL(GL,DF,XLAM,ICR,Y,N)
0044 ISN 0044 RETURN
0045 ISN 0045 END
0046 ISN 0046
0047 ISN 0047
0048 ISN 0048
0049 ISN 0049
0050 ISN 0050
0051 ISN 0051
0052 ISN 0052
0053 ISN 0053
0054 ISN 0054

```


LEVEL 21.8 (JUN 74)

OS/360 FORTRAN H

DATE

COMPILER OPTIONS = NAME= MAIN,OPT=00,LINECNT=60,SIZE=0000K,
 REAL FUNCTION SUM*8 (X,N)
 IMPLICIT REAL*8 (A-H,O-Z)

C
 C
 C

PERFORM UTILITY FUNCTIONS

```

ISN 0004 REAL*8 X(N),Y(N)
ISN 0005 REAL*8 S,SUM2
ISN 0006 S=0.D0
ISN 0007 DO 20 I=1,N
ISN 0008 S=S+X(I)
ISN 0009 SUM=S
ISN 0010 RETURN
ISN 0011 ENTRY SUM2(X,N)
ISN 0012 S=0.D0
ISN 0013 DO 30 I=1,N
ISN 0014 S=S+X(I)*X(I)
ISN 0015 SUM2=S
ISN 0016 RETURN
ISN 0017 ENTRY DOT(X,Y,N)
ISN 0018 S=0.D0
ISN 0019 DO 40 I=1,N
ISN 0020 S=S+X(I)*Y(I)
ISN 0021 DOT=S
ISN 0022 RETURN
ISN 0023 END
  
```

```

COMPILER OPTIONS = NAME= MAIN,OPT=00,LINECNT=60,SIZE=0000K,
SOURCE,EBCDIC,NOLIST,NODECK,LOAD,MAP,NOEDIT,ID,XREF
SUBROUTINE MCHLSK(KK,NV,DUM)
IMPLICIT REAL*8 (A-H,O-Z)

```

ISN 0002
ISN 0003

DVR04770
DVR04780

THIS ROUTINE COMPUTES THE MODIFIED CHOLESKY DECOMPOSITION (MCD) OF A SYMMETRIC MATRIX STORED IN SYM. STORAGE MODE. THE DECOMPOSITION OVERLAYS THE ELEMENTS OF THE MATRIX. IF THE MATRIX IS NOT POSITIVE DEFINITE, THE MCD REPRESENTS A MODIFIED MATRIX WHICH IS. THIS IS DONE BY SETTING THE APPROPRIATE ELEMENTS OF D TO 1.

SEE E.G., "THE USE OF THE MODIFIED CHOLESKY DECOMPOSITION IN DIVERGENCE AND CLASSIFICATION CALCULATIONS", BY D.L. VAN ROOY, M.S. LYNN, & C.H. SNUDER, ICSA TECH. RPT. 275-025-008, RICE UNIV., HOUSTON, TX, MAY, 1973.

[illegible]

KK - THE COVARIANCE MATRIX STORED IN SYMMETRIC STORAGE MODE.

KK=L DL*

WHERE L IS UNIT LOWER TRIANGULAR & D DIAGONAL WITH POS. ENTRIES.

NV - THE NUMBER OF CHANNELS USED

DUM - A DOUBLE PRECISION WORK AREA OF SIZE NV-1

```
REAL*8 KK(1)
REAL*8 DUM(1)
REAL*8 R,R1,T1,TF
LOGICAL*1 JE1
JE1=.TRUE.
EPS=.1
J1=0
JD=0
```

ISN	0004
ISN	0005
ISN	0006
ISN	0007
ISN	0008
ISN	0009
ISN	0010
ISN	0011

LOOP OVER ALL DIMENSIONS

```
DO 10 J=I,NV
  KL=J-1
  L=J+1
  JD=J1
  J1=J1+J
  TF=0.D0
  IF (JEL) GO
  K1=0
```

ISN	0012
ISN	0013
ISN	0014
ISN	0015
ISN	0016
ISN	0017
ISN	0018
ISN	0020

```

COMPUTE THE DIAGONAL ELEMENTS OF D AND STORE IN KK
TEMPORARILY STORE THE PRODUCT KK(I,I)*KK(J,I) IN DUM(I)

```

```
DO 15 I=1,KL
  R=K(K(JD+I))
  K1=K1+I
  R1=K(K(K1)*R)
  TF=TF-R1*R
  DUM(I)=R1
  CONTINUE
CONTINUE
TF=TF+K(K(J1))
```

ISN	0021
ISN	0022
ISN	0023
ISN	0024
ISN	0025
ISN	0026
ISN	0027
ISN	0028
ISN	0029

(14)

U

ISN	0030
ISN	0032
ISN	0033
ISN	0035
ISN	0036
ISN	0037
ISN	0038
ISN	0039
ISN	0041
ISN	0042
ISN	0043
ISN	0044
ISN	0045
ISN	0046
ISN	0047
ISN	0048
ISN	0049

DVR05180
DVR05210
DVR05220
DVR05230
DVR05240
DVR05250
DVR05260
DVR05270

DVR05290
DVR05300
DVR05310
DVR05320

DVR05340
DVR05350
DVR05360

DVR05460

```

COMPILER OPTIONS = NAME= MAIN,OPT=00,LINECNT=60,SIZE=0000K,
SOURCE,EBCDIC,NOLIST,NODECK,LOAD,MAP,NODEJIT,ID,XREF
ISN 0002 SUBROUTINE PDSOLV(A,X,Y,N)
ISN 0003 IMPLICIT REAL*8 (A-H,O-Z)
ISN 0004 REAL*8 A(1),X(N),Y(N)
ISN 0005 REAL*8 S
C
C SOLVE A*X=Y WHERE A IS A SYMMETRICALLY STORED MATRIX CONTAINING
C THE MODIFIED CHOLESKY DECOMPOSITION OF ANOTHER MATRIX.
C
ISN 0006 ISB(J,I)=J*(J-1)/2+1
C
C SOLVE L*Z=Y & PUT RESULT IN X
C
C K=0
C DO 10 I=1,N
C S=Y(I)
C IF (I.EQ.1) GO TO 15
C I1=I-1
C DO 20 J=1,I1
C K=K+1
C 20 S=S-A(K)*X(J)
C 15 K=K+1
C 10 X(I)=S
C
C SOLVE D*B=Z FOLLOWED BY L=TRANSPOSE * X=B & OVERWRITE RESULT ON X
C
C LL=N*(N+1)/2
C X(N)=X(N)/A(LL)
C IF (N.LE.1) RETURN
C DO 30 I1=2,N
C I=N-I1+1
C LL=LL-I+1
C X(I)=X(I)/A(LL)
C I1=I+1
C S=X(I)
C DO 40 J=I1,N
C L=ISB(J,I)
C 40 S=S-A(L)*X(J)
C 30 X(I)=S
C RETURN
C END
ISN 0018
ISN 0019
ISN 0020
ISN 0021
ISN 0022
ISN 0023
ISN 0024
ISN 0025
ISN 0026
ISN 0027
ISN 0028
ISN 0029
ISN 0030
ISN 0031
ISN 0032
ISN 0033

```

```

COMPILER OPTIONS = NAME= MAIN,OPT=00,LINECNT=60,SIZE=0000K,
SOURCE,EBCDIC,NOLIST,NODECK,LOAD,MAP,NODEIT,ID,XREF
SUBROUTINE COMPT (LD,T,Z,N, V,TMP)
IMPLICIT REAL*8 (A-H,O-Z)

C
C THIS ROUTINE IS AN IMPLEMENTATION OF THE COMPOSITE - T ALGORITHM
C TO PERFORM A RANK 1 UPDATE OF THE MCD STORED IN ARRAY LD(I,E.
C K=L*D*L-TRANS & WE WISH TO COMPUTE L' &D' S.T.K'=K+Z*Z-TRANS/T(I)
C & K'=L'*D'*L'-TRANS).
C THE COMPOSITE-T ALGORITHM WAS DEVELOPED BY FLETCHER & POWELL
C SEE 'ON THE COMPUTATION & UPDATING OF THE MODIFIED CHOLESKY
C DECOMPOSITION OF A COVARIANCE MATRIX,' BY D.L. VAN ROOY, ICSA
C TECH.RPT. 275-025-024,RICE UNIV., HOUSTON, TX. MAY, 1976.
C
ISN 0004 REAL*8 LD(1),Z(N)
ISN 0005 REAL*8 T(1), V(N)
ISN 0006 REAL*8 TMP(N)
ISN 0007 REAL*8 S
ISN 0008 LOGICAL*1 TPOS,RNDERR,LALP

C LD = ARRAY CONTAINING L & D STORED IN SYM.STORAGE MODE
C T = AN N+1 VECTOR WHOSE FIRST ELEMENT IS AS ABOVE
C Z = VECTOR OF THE UPDAEE AS ABOVE
C N = THE DIMENSION
C V = WORKING STORAGE OF LENGTH .GE. N
C TMP = DOUBLE PRECISION WORKING STORAGE OF LENGTH .GE. N

ISN 0009 TPOS=T(1)*GT.0.
ISN 0010 IF (TPOS) GO TO 35

C A POINT IS TO BE DELETED

ISN 0012 EPS=5.97E-8

C SOLVE L*V=Z FOR V

ISN 0013 K=1
ISN 0014 V(1)=Z(1)
ISN 0015 DO 10 I=2,N
ISN 0016 IJ=I-1
ISN 0017 S=0.00
ISN 0018 DO 15 J=1,IJ
ISN 0019 K=K+1
ISN 0020 S=S+LD(K)*V(J)
ISN 0021 K=K+1
ISN 0022 V(I)=Z(I)-S
ISN 0023 10 CONTINUE
15 S=S+LD(K)*V(J)
K=K+1
V(I)=Z(I)-S
10 CONTINUE

C COMPUTE THE T(I*S)

C
ISN 0024 K=0
ISN 0025 RNDERR=.FALSE.
ISN 0026 DO 20 I=1,N
ISN 0027 K=K+1
ISN 0028 TMP(I)=V(I)*V(I)/LD(K)
ISN 0029 T(I+1)=T(I)+TMP(I)
ISN 0030 IF (T(I+1).GE.0.) RNDERR=.TRUE.
ISN 0032 20 CONTINUE

```



```

ISN 0033      IF (.NOT.RNDERR) GO TO 35
C             ROUNDING ERROR HAS MADE A  $T(I+1) \geq 0$ , SO CORRECT FOR THIS
C
ISN 0035      T(N+1)=EPS*T(I)
ISN 0036      DO 30 J=1,N
ISN 0037      I=N-J+1
ISN 0038      T(I)=T(I+1)-TMP(I)
ISN 0039      30 CONTINUE
ISN 0040      35 CONTINUE
ISN 0041      IJ=0
ISN 0042      DO 40 I=1,N
ISN 0043      II=I+1
ISN 0044      IJ=IJ+I
ISN 0045      V(I)=Z(I)
ISN 0046      DI=LD(IJ)
ISN 0047      IF (DI.GT.0.) GO TO 44
C
C             D(I) = 0, SO RANK OF D WILL EITHER INCREASE OR REMAIN UNCHANGED.
C
ISN 0049      IF (DABS(V(I)).GT.1.E-30) GO TO 42
C             RANK OF D WILL REMAIN UNCHANGED
C
ISN 0051      T(I+1)=T(I)
ISN 0052      GO TO 40
C
C             RANK OF D WILL INCREASE BY 1
C
ISN 0053      42 LD(IJ)=V(I)*V(I)/T(I)
ISN 0054      IF (I.EQ.N) RETURN
ISN 0055      K=IJ
ISN 0056      DO 45 J=II,N
ISN 0057      K=K+J-1
ISN 0058      LD(K)=Z(J)/V(I)
ISN 0059      45 CONTINUE
ISN 0060      RETURN
ISN 0061      44 CONTINUE
ISN 0062
C             UPDATE D
C
ISN 0063      IF (TPOS) T(I+1)=T(I)+V(I)*V(I)/DI
ISN 0065      ALP=T(I+1)/T(I)
ISN 0066      LD(IJ)=DI*ALP
ISN 0067      IF (I.EQ.N) RETURN
C
C             UPDATE L & MODIFY Z ACCORDINGLY
C
ISN 0069      BETA=(V(I)/DI)/T(I+1)
ISN 0070      LALP=.FALSE.
ISN 0071      IF (ALP.LE.4.) GO TO 52
C
C             THIS METHOD USED TO INSURE STABILITY IF ALPHA GT. 4
C
ISN 0073      LALP=.TRUE.
ISN 0074      GAM=T(I)/T(I+1)
ISN 0075      K=IJ
ISN 0076      DO 50 J=II,N
ISN 0077      K=K+J-1

```

```

ISN 0078
ISN 0079
ISN 0080
ISN 0081
ISN 0082
ISN 0083
ISN 0084
ISN 0085
ISN 0086
ISN 0087
ISN 0088
ISN 0089
ISN 0090
ISN 0091

XX=GAM*LD(K)+BETA*Z(J)
Z(J)=Z(J)-V(I)*LD(K)
LD(K)=XX
50 CONTINUE
GO TO 40
52 K=IJ
DO 60 J=I1,N
K=K+J-1
Z(J)=Z(J)-V(I)*LD(K)
LD(K)=LD(K)+BETA*Z(J)
60 CONTINUE
40 CONTINUE
RETURN
END

```

```

COMPILER OPTIONS = NAME= MAIN,OPT=00,LINECNT=60,SIZE=0000K,
SOURCE,EBCDIC,NOLIST,NODECK,LOAD,MAP,NODEIT,ID,XREF
SUBROUTINE LDMUL ( H,S,R,T,N)
IMPLICIT REAL*8 (A-H,O-Z)
C
C MULTIPLY THE MATRIX WHOSE MCD IS STORED IN H (IN SYM. STORAGE
C MODE) BY S & STORE THE RESULT IN R
C T IS WORKING STORAGE OF LENGTH .GE.N-1
C
      REAL*8 H(1),S(N),R(N),T(N)
      REAL*8 SK
      II=1
      DO 10 I=1,N
C
C      MULTIPLY BY L-TRANPOSE
C
C
C      IS=II
C      SK= S(I)
C      IF (I.GE.N) GO TO 25
C      IP1=I+1
C      DO 20 J=IP1,N
C      IS=IS+J-1
C      SK=SK+S(J)*H(IS)
C      20 SK=SK+S(J)*H(IS)
C
C      MULTIPLY BY D
C
C      25 SK=SK*H(II)
C      T(I)=SK
C
C      MULTIPLY BY L
C
C      IF (I.LE.1) GO TO 35
C      IM1=I-1
C      IJ=II-1
C      DO 30 J=1,IM1
C      IJ=IJ+1
C      SK=SK+T(J)*H(IJ)
C      30 SK=SK+T(J)*H(IJ)
C      35 R(I)=SK
C      II=II+1
C      10 CONTINUE
C      RETURN
C      END

```